AI Planning

Exercise Sheet 2

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Exercise 2.1

(a) Transform the operator

```
original  \langle \neg e \lor f, (a \rhd (b \rhd c)) \land (\neg d \rhd c) \land (\neg (\neg c \land \neg a) \rhd (d \land \neg d)) \land (d \rhd \neg e) \rangle  (7)  \langle \neg e \lor f, ((a \land b) \rhd c) \land (\neg d \rhd c) \land (\neg (\neg c \land \neg a) \rhd (d \land \neg d)) \land (d \rhd \neg e) \rangle  (9)  \langle \neg e \lor f, (((a \land b) \lor \neg d) \rhd c) \land (\neg (\neg c \land \neg a) \rhd (d \land \neg d)) \land (d \rhd \neg e) \rangle  (8)  \langle \neg e \lor f, (((a \land b) \lor \neg d) \rhd c) \land (\neg (\neg c \land \neg a) \rhd d) \land (\neg (\neg c \land \neg a) \rhd \neg e) \land (d \rhd \neg e) \rangle  (9)  \langle \neg e \lor f, (((a \land b) \lor \neg d) \rhd c) \land (\neg (\neg c \land \neg a) \rhd d) \land ((\neg (\neg c \land \neg a) \lor d) \rhd \neg e) \rangle  2x deMorgan  \langle \neg e \lor f, (((a \land b) \lor \neg d) \rhd c) \land ((c \lor a) \rhd d) \land ((c \lor a \lor d) \rhd \neg e) \rangle
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(b) Transform the ENF operator

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original (\neg e \lor f, (((a \land b) \lor \neg d) \triangleright c) \land ((c \lor a) \triangleright d) \land ((c \lor a \lor d) \triangleright \neg e))
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Exercise 2.2

(a-c)

Listing 1: set cover problem as a PDDL domain

Listing 2: set cover instance as a PDDL problem

Listing 3: solving set cover instance with fast-downward

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$ ./fast-downward.py scprob.pddl --search "astar(blind())"
select-set s4 (1)
select-set s5 (1)
Plan length: 2 step(s).
Plan cost: 2
Initial state h value: 1.
Expanded 6 state(s).
Reopened O state(s).
Evaluated 16 state(s).
Evaluations: 16
Generated 21 state(s).
Dead ends: 0 state(s).
Expanded until last jump: 1 state(s).
Reopened until last jump: 0 state(s).
Evaluated until last jump: 6 state(s).
Generated until last jump: 5 state(s).
Number of registered states: 16
Search time: Os
Total time: Os
Solution found.
Peak memory: 2924 KB
$ cat sas_plan
(select-set s4)
(select-set s5)
; cost = 2 (unit cost)
```

(d)

- Optimal planning: every possible plan is considered. It is guaranteed that the optimal plan (lowest cost) can be identified.
- Satisficing planning: it is sufficient to identify a plan that leads to the goal.
- Arbitrary set covers: as long as every element is covered we're satisfied.
- Cardinality minimal set covers: the goal is to pick $\mathcal{C} \subseteq \mathcal{S}$ such that every element is covered and $|\mathcal{C}|$ is minimal.

Since our formalization does only define complete coverage of elements as the goal, cardinal minimality can only be achieved by applying optimal planning. Since fast-downward increments the cost of a plan for each applied operation and the only operator defined is selecting an element of S, cost of a plan equals to |C|. Finding the

optimal plan therefore means finding the minimal $\mathcal{C}\subseteq\mathcal{S}$ such that every element is covered.

In our formalization plan existence is equivalent to the existence of a set cover. If we were to include an element $e_5 \in \mathcal{U}$ such that $\forall S_i \in \mathcal{S} : e_5 \notin S_i$ there would be no possible set cover and therefore no existing plan that solves the problem.